

# Computational complexity

...

Evaluating algorithms

# An intuitive approach to complexity

Search algorithm in a sorted array :

Target value  $T$

Sorted array  $A = \{A_1, A_2, A_3 \dots A_n\}$

Search for 47								
0	4	7	10	14	23	45	47	53

Goal: find an algorithm to search if  $T$  is present in  $A$

# An intuitive approach to complexity

Two algorithms:

## Linear search algorithm

```
trouvé = False
i = 0
while not trouvé and i < len(A) :
    if(A[i] == T):
        trouvé = True
    i += 1

if trouvé :
    print("trouvé")
else:
    print("pas trouvé")
```

## Binary search algorithm

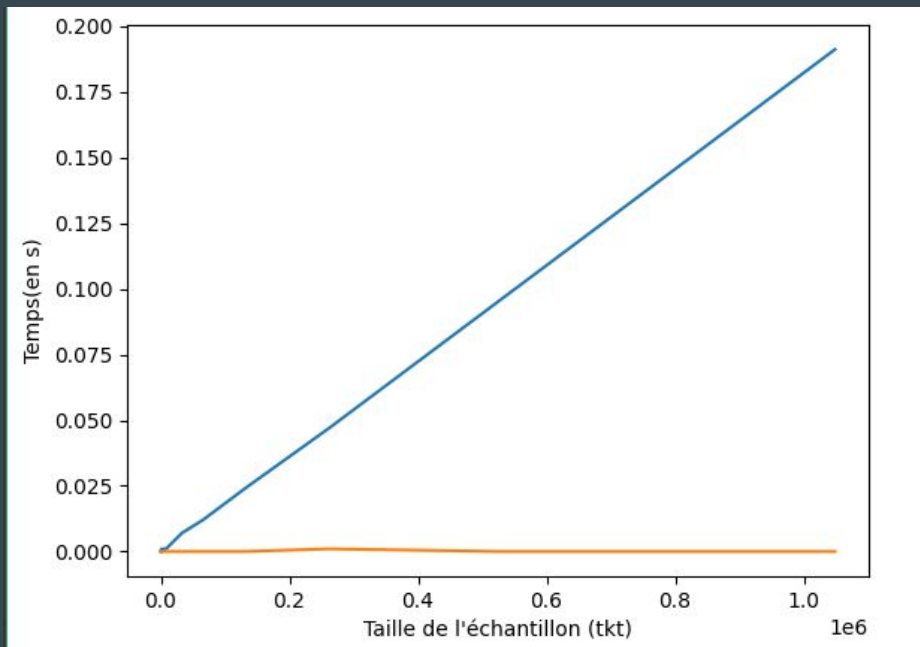
```
start = 0
end = len(A) - 1
while(start <= end):
    mid = (start + end) // 2
    if(A[mid] > T):
        end = mid - 1
    elif(A[mid] < T):
        start = mid + 1
    else:
        print("trouvé")

print("pas trouvé")
```

How efficient are they?

# An intuitive approach to complexity

Let's compare them with the module *time*:



In blue, the Linear search algorithm

In orange, the Binary search algorithm

*What happened?*

# An intuitive approach to complexity

## Linear search algorithm

```
trouvé = False
i = 0
while not trouvé and i < len(A) :
    if(A[i] == T):
        trouvé = True
    i += 1

if trouvé :
    print("trouvé")
else:
    print("pas trouvé")
```

Loop of size  $n$  at worst, with fixed operations inside  $\rightarrow$  complexity is called *linear* in  $n$

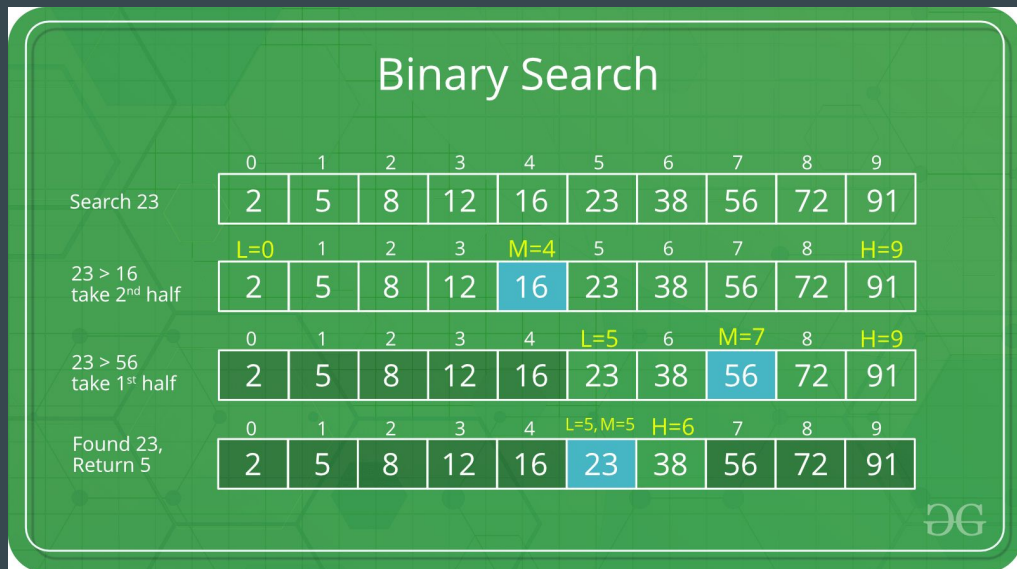
## Binary search algorithm

```
start = 0
end = len(A) - 1
while(start <= end):
    mid = (start + end) // 2
    if(A[mid] > T):
        end = mid - 1
    elif(A[mid] < T):
        start = mid + 1
    else:
        print("trouvé")

print("pas trouvé")
```

Let's take a closer look at what happens with this algorithm...

# An intuitive approach to complexity

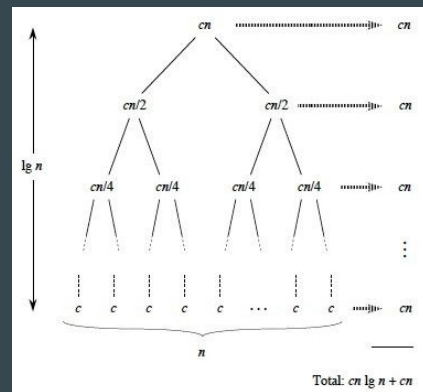


Binary search algorithm

```
def fib(n):  
    if n <= 1:  
        return n  
    else:  
        return fib(n-1) + fib(n-2)
```

```
n = int(input())  
print(fib(n))
```

At each step the search interval is divided in half.  
→ complexity is called *logarithmic* in  $\log_2(n)$ .



# A more formal definition

- What we have evaluated previously is the *time complexity*, i.e the amount of time that it takes to run an algorithm, in function of its parameters
- When comparing memory consumption, we talk about *space complexity*

# A more formal definition

When evaluating complexity, we use the *big O notation*:

$$f(x) = O(g(x))$$

*if and only if* there is  $x_0$  and  $M$  such that:

$$|f(x)| \leq M g(x) \quad \text{for all } x > x_0$$

For example:

- $n^2 - 3n + 5 = O(n^2)$
- $42 = O(1)$
- $-7 n \log_2 n = O(n \log n)$

This notation allows to evaluate how the program behaves depending on the size of the parameters.



# Examples of time complexity

Let  $n$  be the size of an array.

Here are the time complexities of a few operations:

- $O(\log n)$  find an element in a sorted array with binary search
- $O(n)$  explore all elements of the array
- $O(n \log n)$  sort the array with a merge sort
- $O(n^2)$  find all the couples of elements of the array
- $O(n!)$  find all the permutations of the array

etc

# Average VS worst case complexity

Sometimes the time needed to run an algorithms varies for inputs of the same size.

For example, the complexity of the quick sort is:

- $O(n \log n)$       on average
- $O(n^2)$             in the worst case

When talking about complexity, we usually refer to the *worst case* complexity.

# Common complexities

- 1 constant
- $\log n$  logarithmic
- $n$  linear
- $n \log n$  linearithmic
- $n^2$  quadratic
- $n^3$  cubic
- $n^k$  polynomial
- $k^n$  exponential
- $n!$  factorial

*Now, some exercises :)*

# Equivalence time - complexity

total time = number of operations / operations per second

*In python3 ...*

total time = number of operations /  $10^7 - 10^8$

# Why is it useful?

Problem's constraints :  $0 \leq N \leq 10^2$ ,  $0 \leq C \leq 10^3$ , Time limit = 1s

$N * C \Rightarrow 10^5$  operations  $\Rightarrow$  0.001s

$N^2 * C \Rightarrow 10^7$  operations  $\Rightarrow$  0.1s

$N * C^2 \Rightarrow 10^8$  operations  $\Rightarrow$  1s

$N! * C \Rightarrow \infty?$  operations  $\Rightarrow$  NO

# Some traps easily avoidable

```
array = ACollectionImplementation()  
targets = [1,2,5,6]  
for element in targets :  
    if(array.found(element)):      ← we don't know how this function is implemented  
        print(element,"was found")  
    else :  
        print(element,"was not found")
```

The Complexity of this algorithm is unknown